



# Bayesian Hierarchical Modelling of Young Stellar Clusters

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# Stellar clusters: Origins



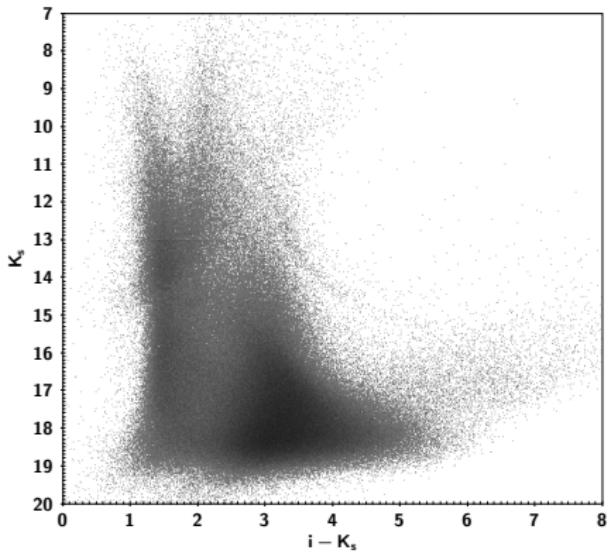
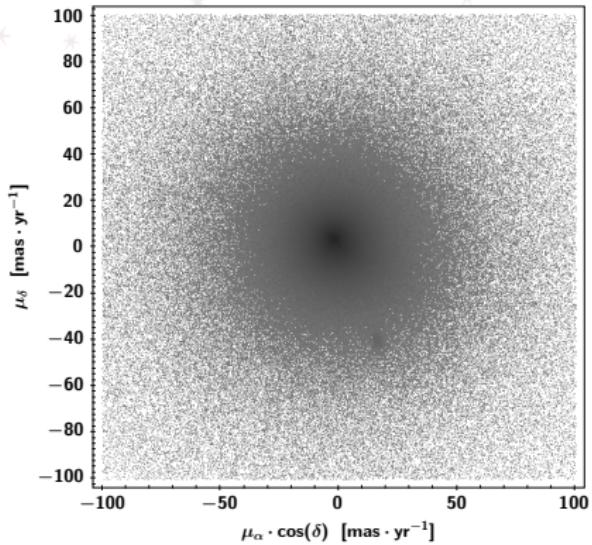
Rogelio Bernal Andreo  
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# Stellar clusters: Origins

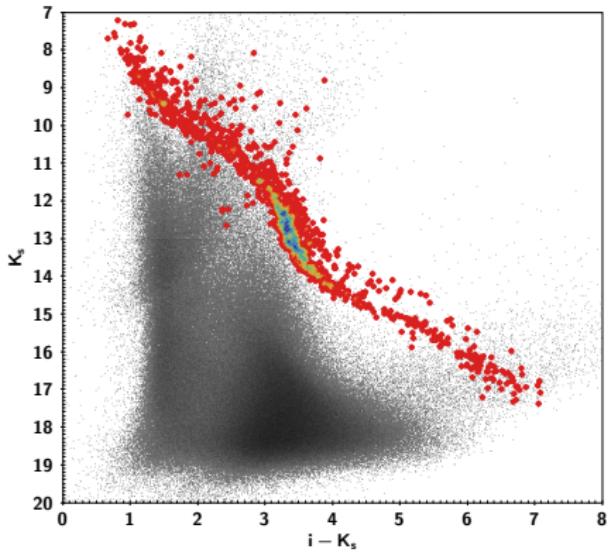
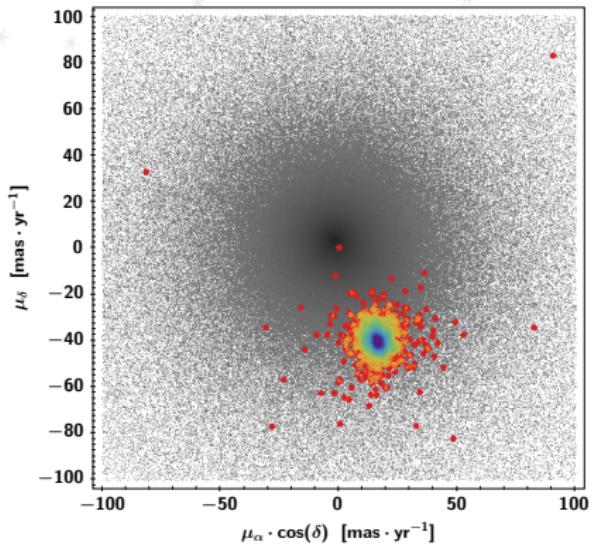


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# Stellar clusters: Observables



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# Objective

## Obtain the statistical distributions of the cluster population

- Sky positions
- Proper Motions
- Apparent magnitudes

Simultaneously:

- Membership probabilities

# Current approach

- BANYAN I (Malo et al. 2013)
- UPMASK (Krone-Martins & Moitinho 2014)
- Sarro (Sarro et al. 2014)
- BANYAN II (Gagné et al. 2015)
- Sampedro (Sampedro & Alfaro 2016)
- LACEwING (Riedel et al. 2017)

## Caveats

- Classifiers only
- Homogeneous uncertainties
- Fully observed objects

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## Remedy

- Cluster distributions
- Heterogeneous uncertainties
- Missing value objects

# Strategy

- 
1. Create a new method.
  2. Benchmark it.
  3. Validate it
  4. Evaluate it.
  5. Discussion of the results.

- Bayesian Hierarchical Model
- Pleiades cluster
- Synthetic data
- DANCe DR2

# Pleiades DANCe DR2 data set

KPNO/Mosaic1

UKIRT/WFCAM

Subaru/SuprimeCam

CFHT/CFHT12K

INT/WFC

CFHT/UH8K

KPNO/NEWFIRM

CTIO/MOSAIC2

CFHT/MegaCam

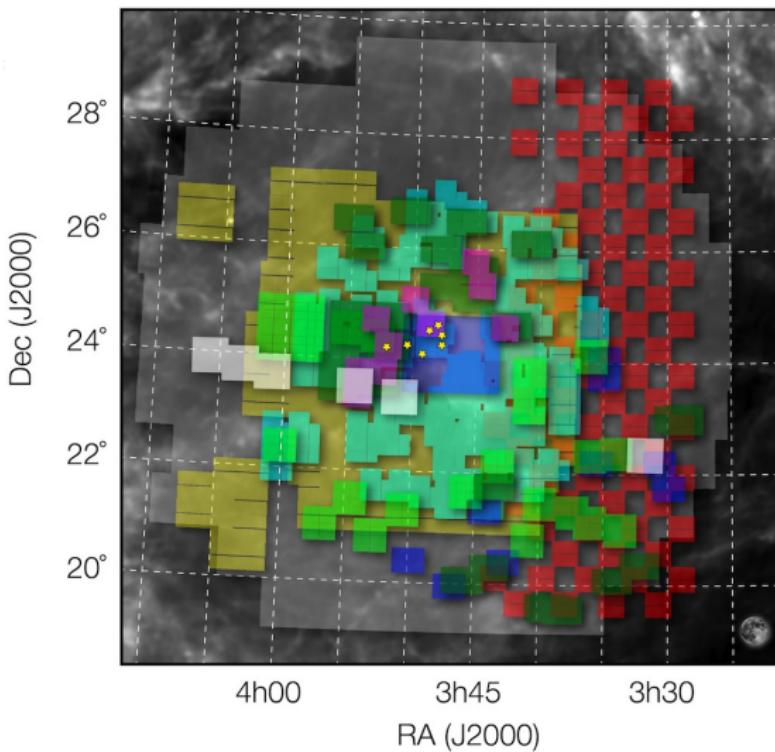
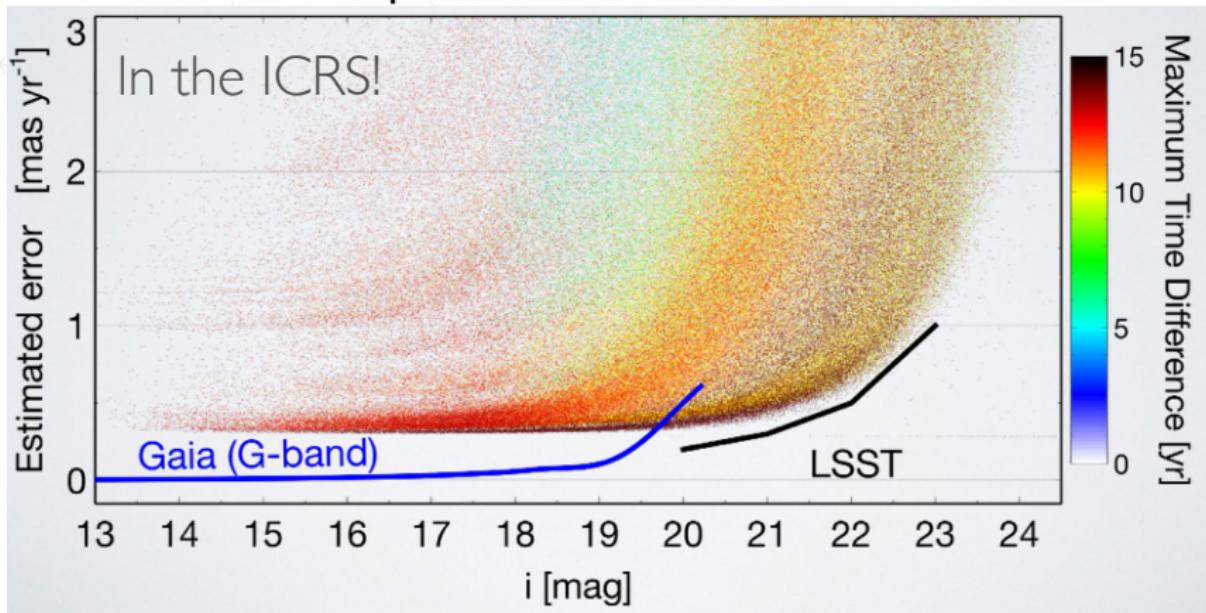


Image by Bouy et al. 2013

# Pleiades DANCe DR2 data set

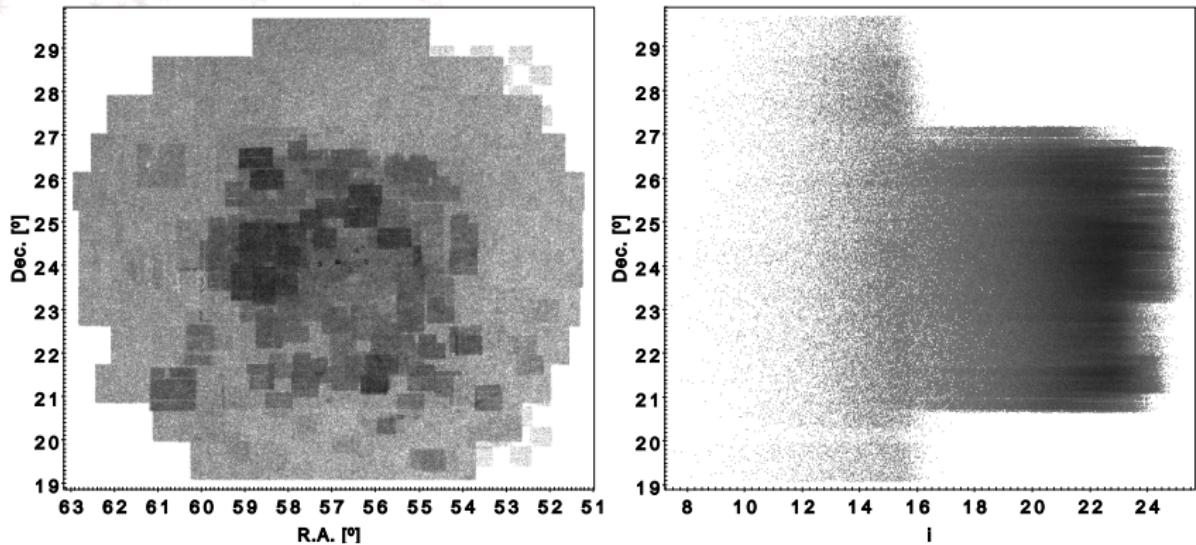
Complement Gaia at the faint end



Credit: Hervé Bouy

# Pleiades DANCe DR2 data set

Complex probability of detection at all scales



# Restricted data set

Reduce data set size

Computational constraints!

## Selection

- Membership probabilities of Bouy et al. 2015
- Size:  $10^5$  objects.
- Probability threshold:  $1 \times 10^{-11}$

## Assumptions

- No cluster member outside the restricted sample

# Overview of the method

## Bayes' theorem

$$p(\boldsymbol{\theta}|\mathbf{D}, \mathcal{M}) = \frac{p(\mathbf{D}|\boldsymbol{\theta}, \mathcal{M}) \cdot p(\boldsymbol{\theta}|\mathcal{M})}{p(\mathbf{D}|\mathcal{M})}$$

## Hierarchical model

Reduce prior subjectivity

## Generative model

True values + Heteroscedastic data

## Missing values

### Types

- Deterministic
- Stochastic

### Simplistic assumptions

- Ignorability
- Missing at random

# Details of the method

## Selection of observables

- Proper motions and  $i - K_s, Y, J, H, K_s$
- Sky positions (independent analysis)

## Generative model

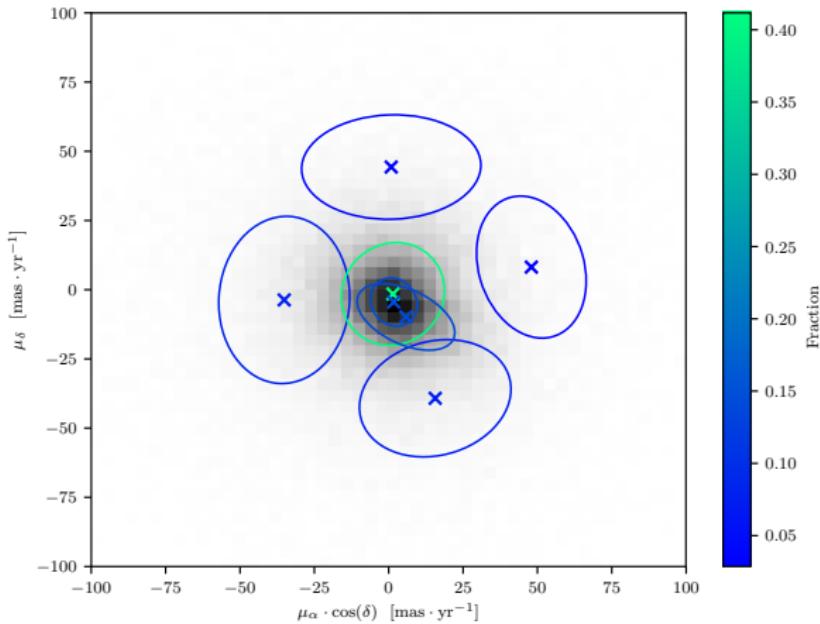
Field + Cluster

## Assumptions

- Observables of Sarro et al. 2014 are correct.
- I hold the field fixed.
- Photometry independent of proper motions.
- Populations: Single stars and Equal-Mass Binaries (EMB).

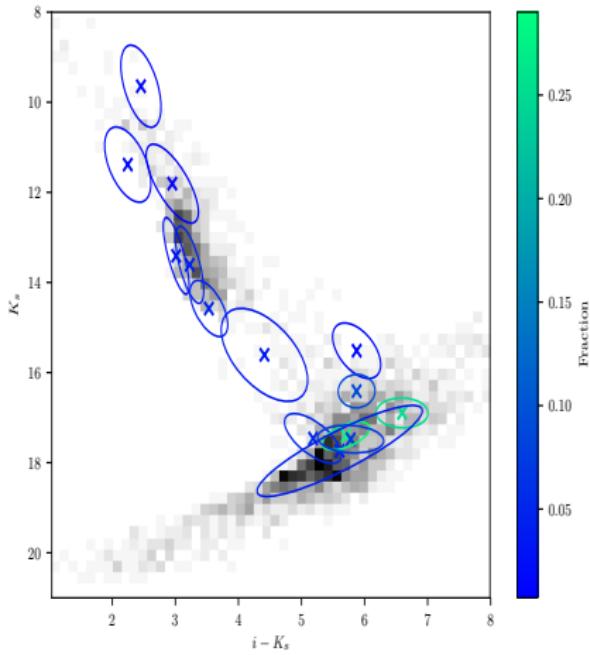
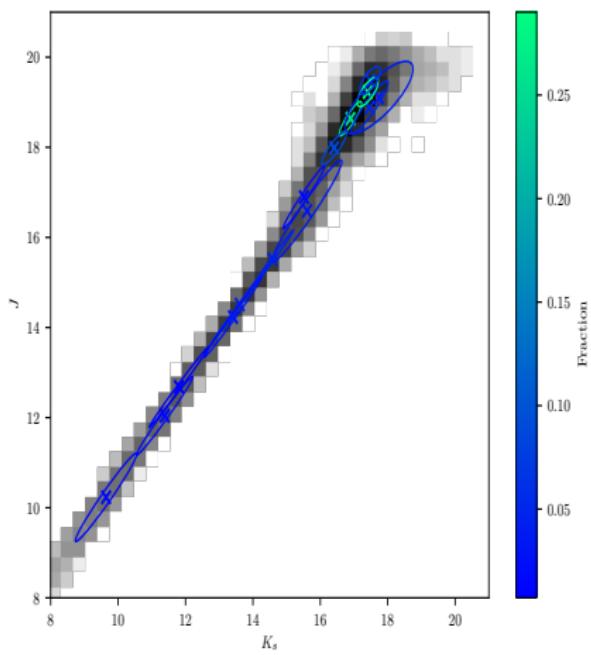
# Field: Proper Motions

Gaussian Mixture Model + Uniform with 8 components (BIC)  
Maximum-Likelihood Estimate



# Field: Photometry

Missing values Gaussian Mixture Model with 14 components (BIC)  
Maximum-Likelihood Estimate



# Cluster model

Single stars + Equal-Mass binaries  $\rightarrow$  85 parameters

## Proper motions

- GMM
  - Singles: 4 components
  - EMB: 2 components

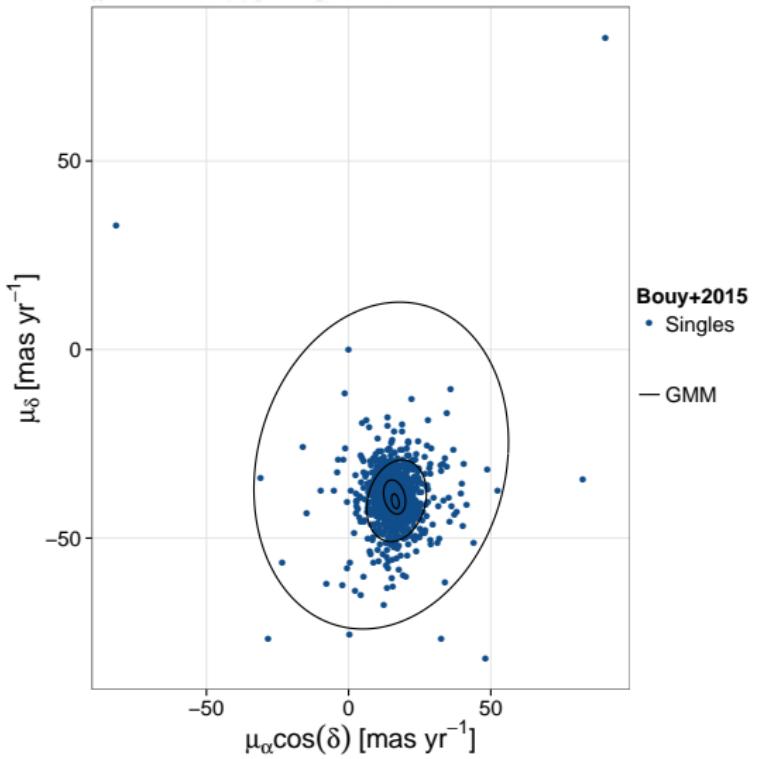
## Photometry

- True  $i - K_s$  distribution
  - GMM: 5 components
- Splines:  $Y, J, H, K_s$
- Intrinsic dispersion
  - 5D Gaussian

## Assumptions

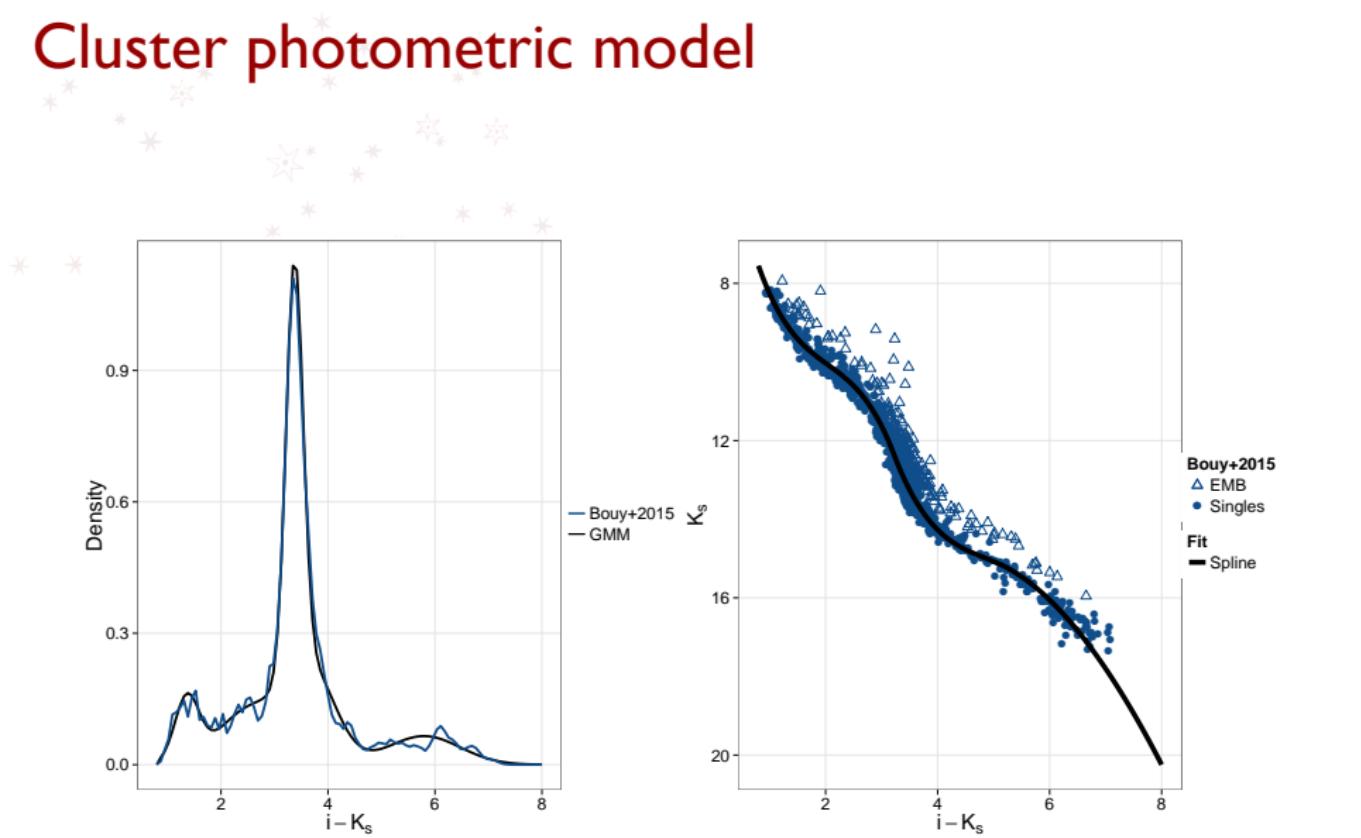
- Number of GMM components given by BIC
- Splines parameter:  $i - K_s$
- Splines knots remain fixed

# Cluster proper motions model



- Gaussian Mixture
- Singles: 4 components
  - EMB: 2 componentes

# Cluster photometric model



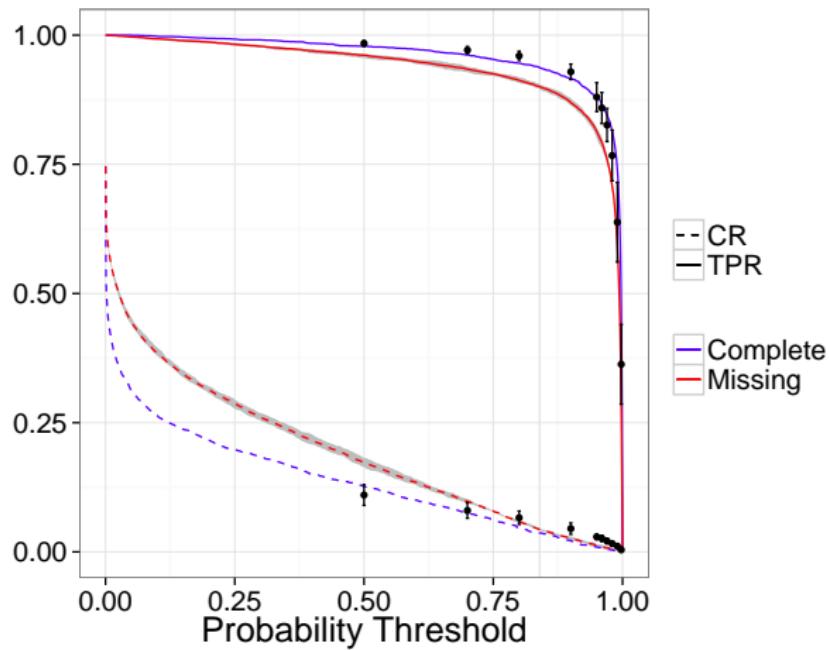
# Obtaining the cluster distributions

## High-Performance Computing

- Diverse infrastructures: UCA,UGA,CAB,UB
- Hybrid MPI-Multithreading approach
- Heuristic Particle Swarm Optimiser
- Markov Chain Monte Carlo
  - Convergence

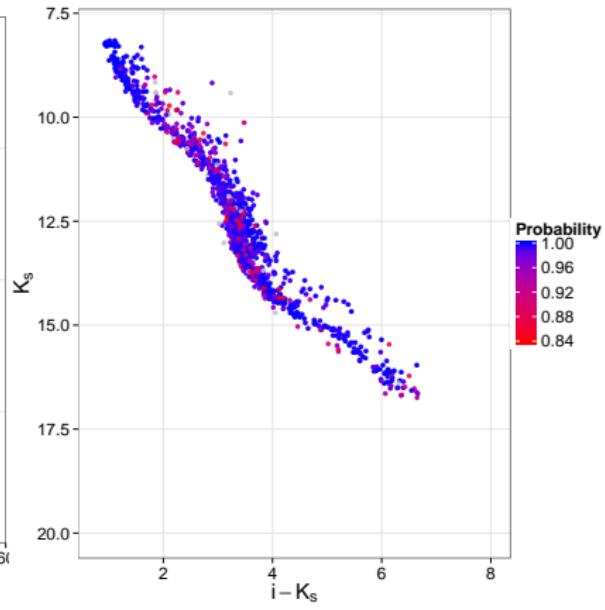
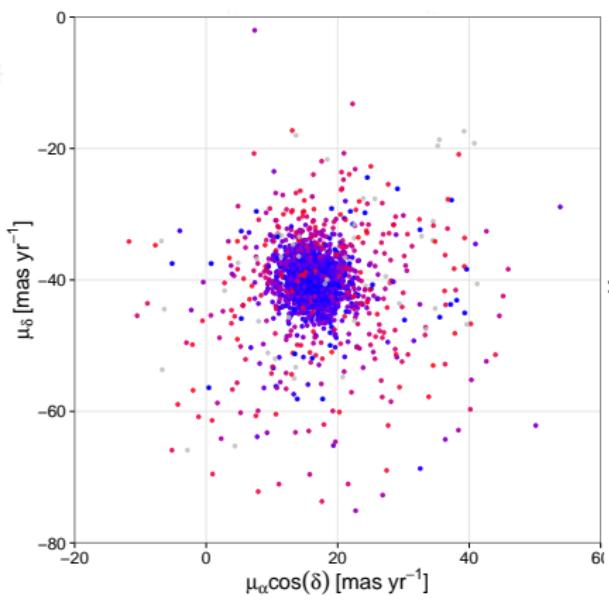
# Validation on synthetic data

- Optimum Threshold = 0.84
- Contamination Rate =  $4.3 \pm 0.2\%$
- True Positive Rate =  $90.0 \pm 0.05\%$
- Expected value of CR =  $5.8 \pm 0.2\%$



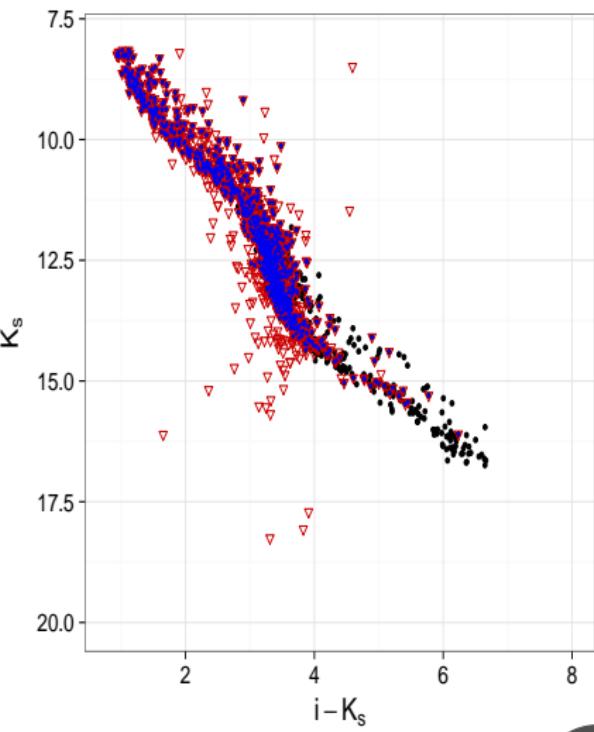
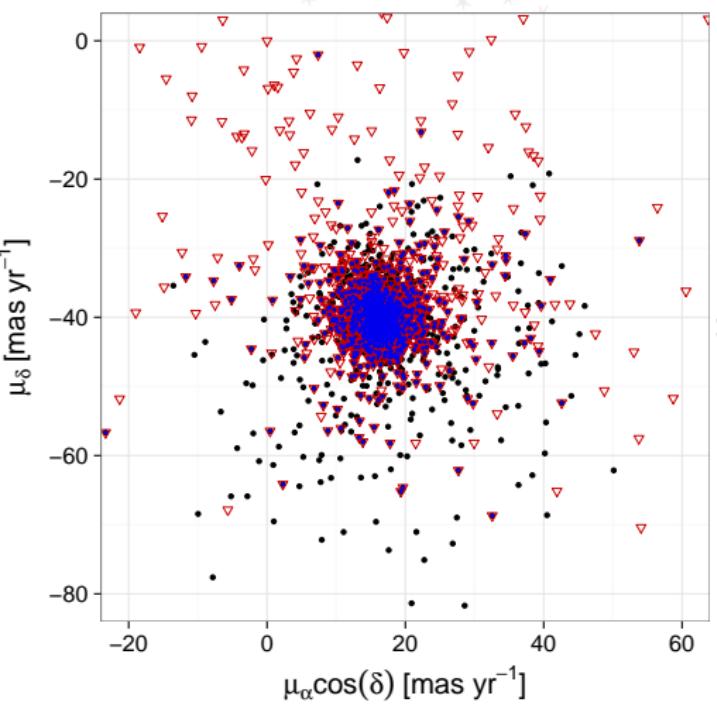
# Results on the DANCe DR2

- Cluster membership probabilities → HMPS



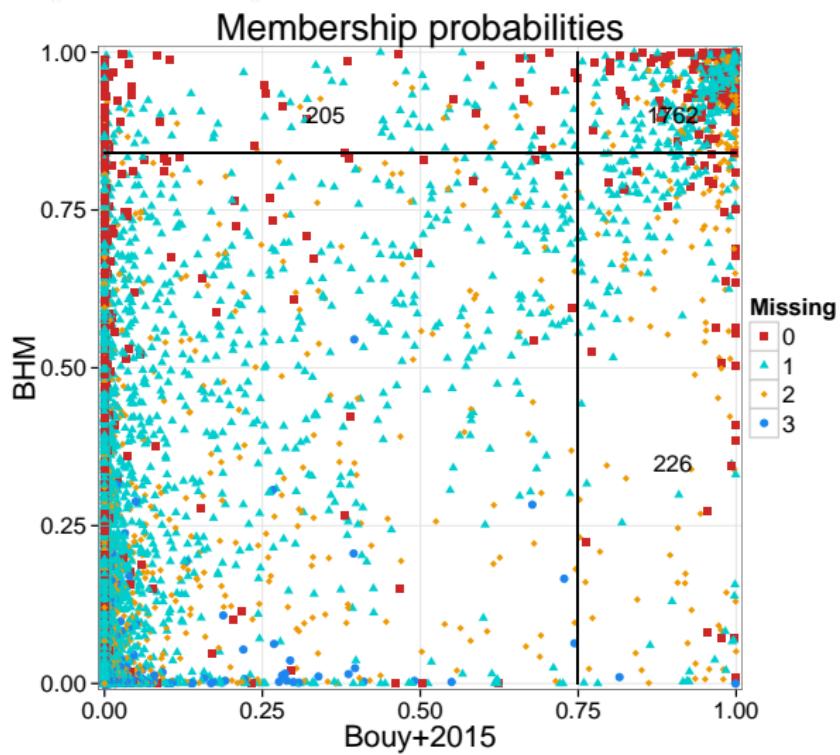
# Evaluation: Stauffer et al. 2007

- Agreement: 80%

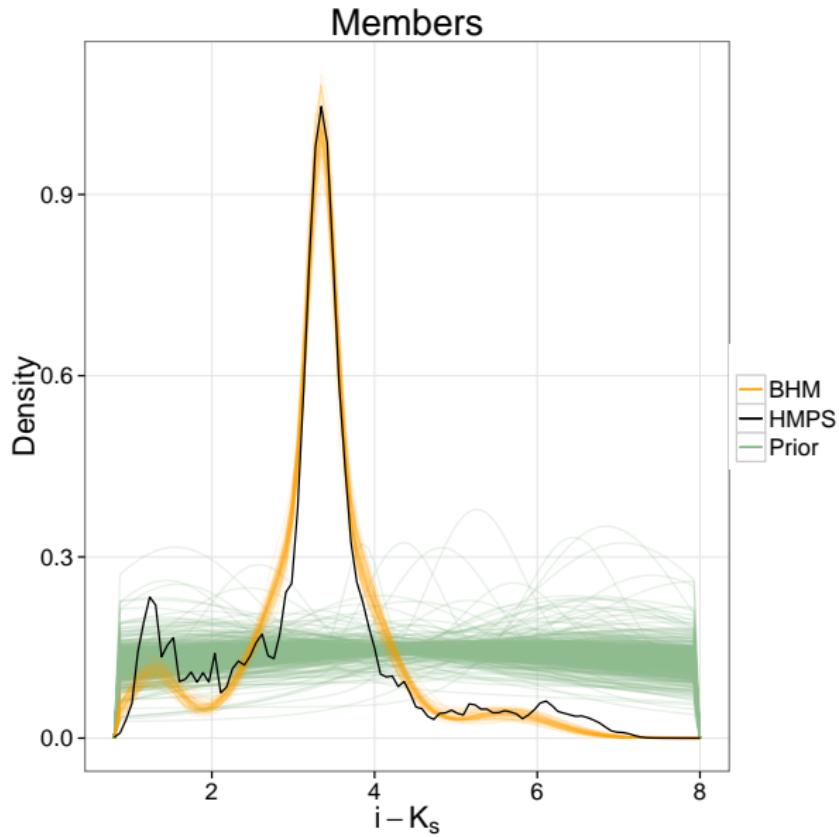


# Evaluation: Bouy et al. 2015

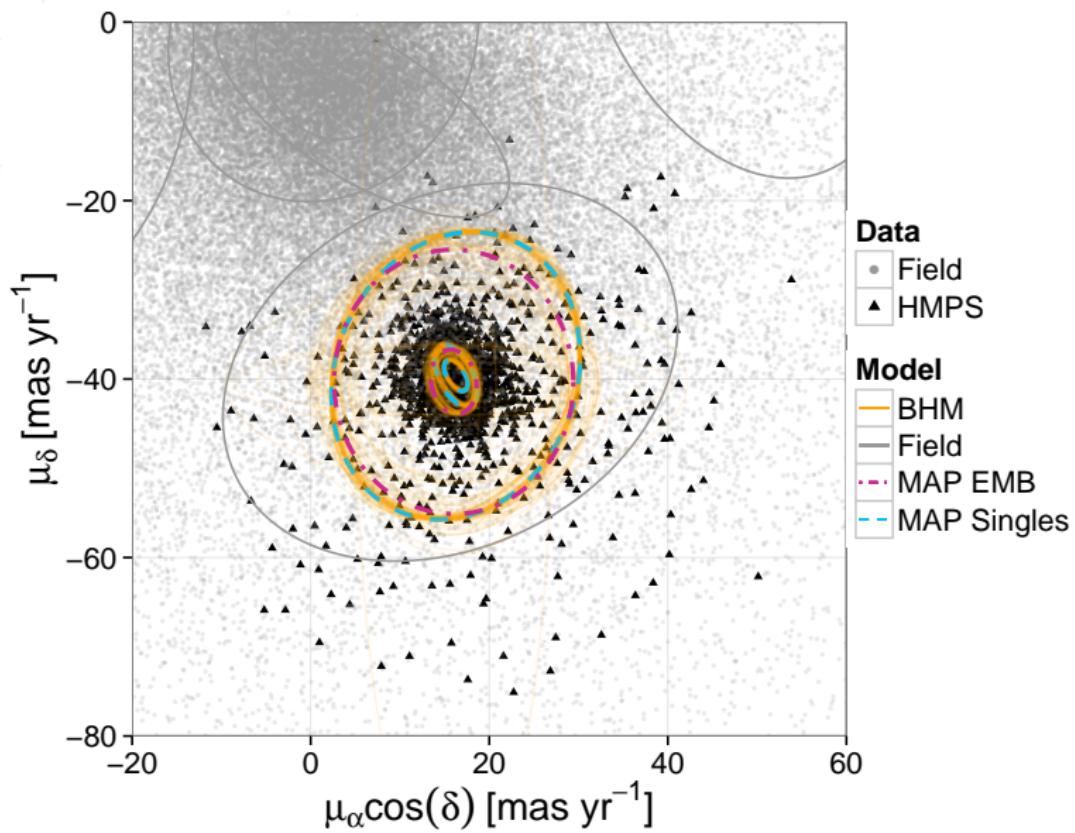
- Agreement: 88%



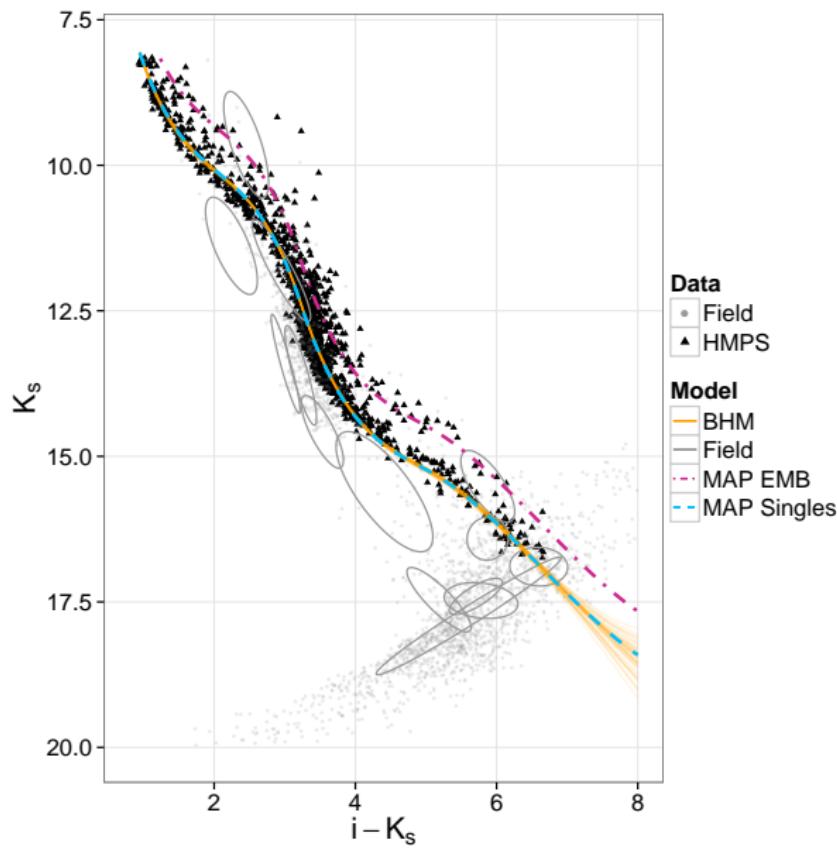
# The statistical distributions: Colour index



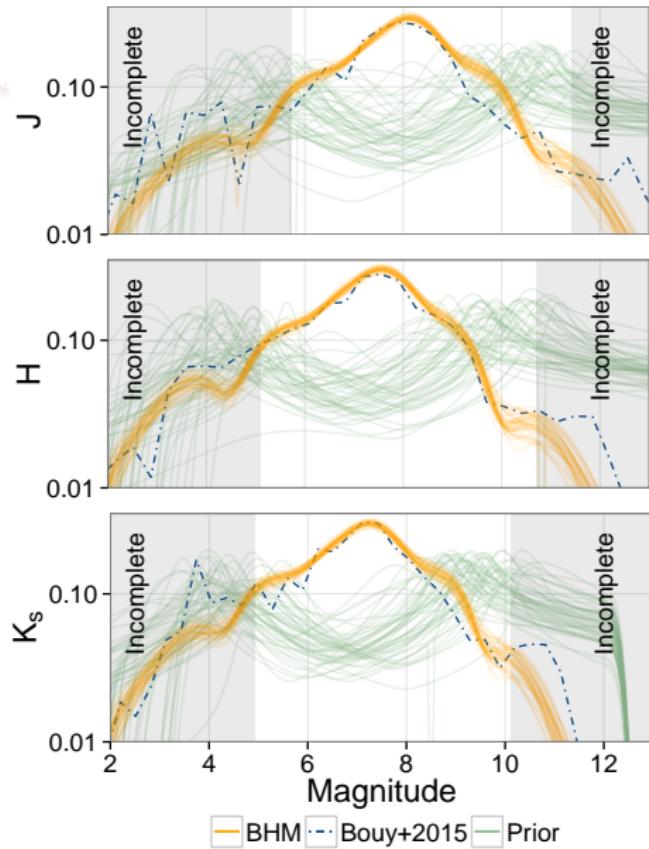
# The statistical distributions: Proper motions



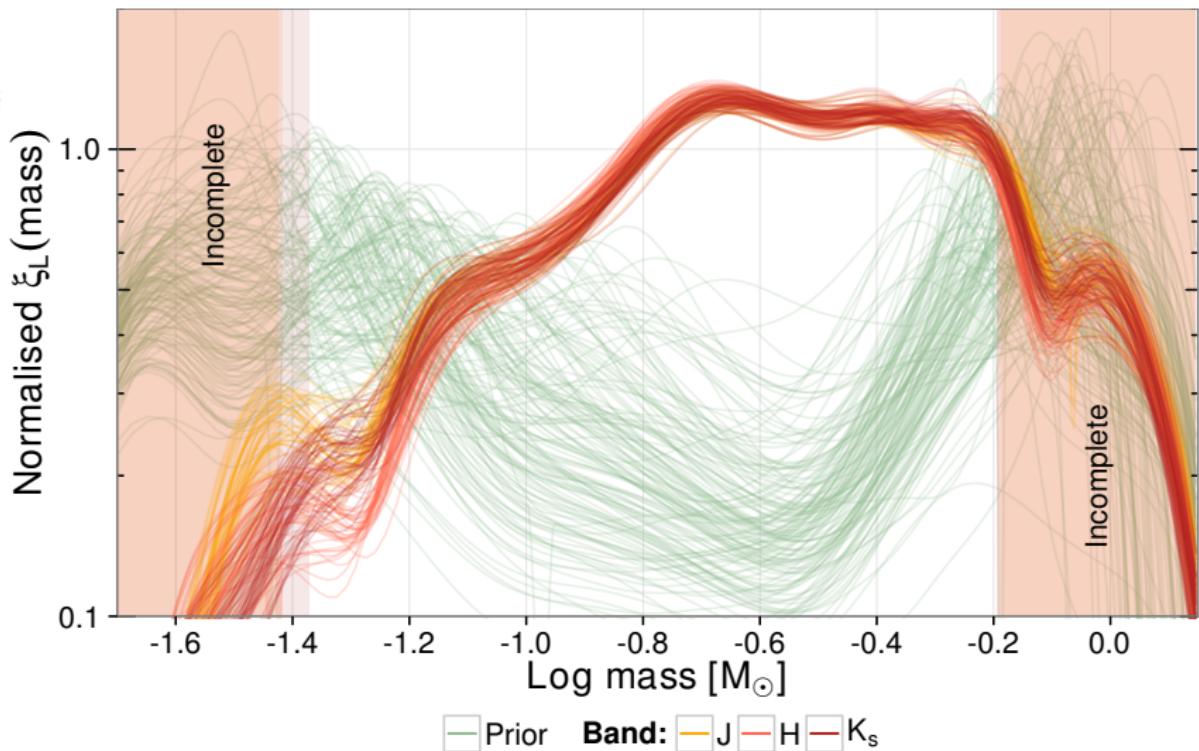
# The statistical distributions: Empirical isochrone



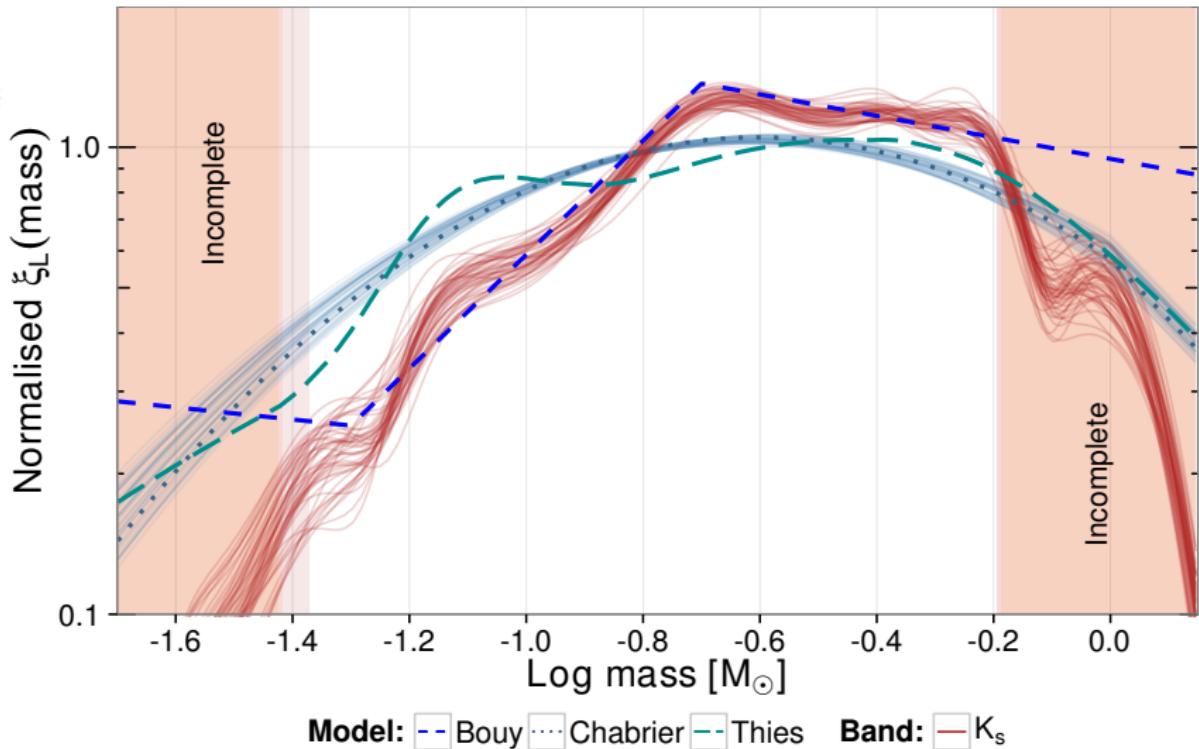
# The statistical distributions: Luminosity



# Results: The mass distribution



# Results: The mass distribution



# The Projected Spatial Distribution

## Bayesian Model Selection

### Models

- King 1962
- Elson et al. 1987
- Lauer et al. 1995
- Generalised King

### Extensions:

- Radially symmetric
- Biaxially symmetric
- Luminosity segregated

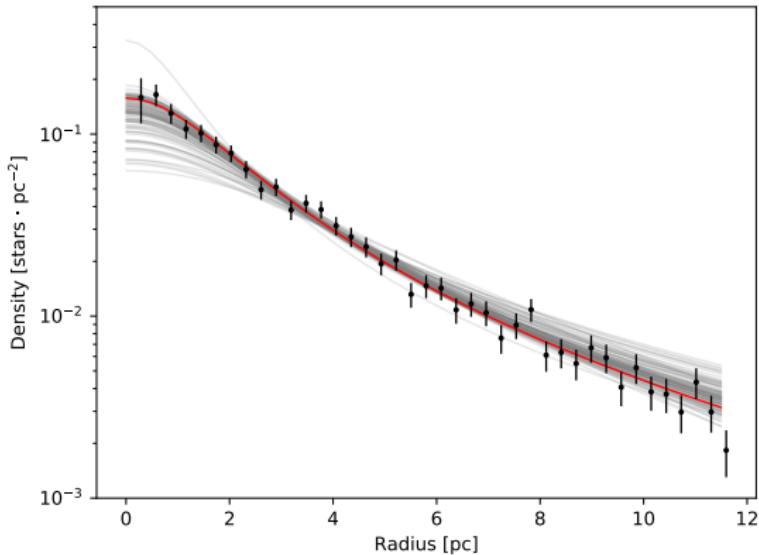
### Data set

- The HMPS

Simultaneously

Bayesian evidence + Parametric inference

# The Projected Spatial Distribution



## Results

- Strong evidence:
  - Ellipticity
  - Luminosity segregation
- Best models
  1. King
  2. Generalised King
- Still lack 20% of members (> 5°)

# Summary of results

- 80 – 90% of candidate members in agreement with the literature
- 205 new cluster candidate members
- Mass distribution in discrepancy with IMFs in low-mass regime
- Projected Spatial Distribution
  - Evidence of ellipticity and mass segregation
  - One fifth of members beyond the survey area
  - Proved the validity of King's family of models

# Conclusions

## Objective fulfilled

- Statistical distributions of the cluster population
  - Sky positions
  - Proper motions
  - Apparent magnitudes → Luminosity → Mass
- Membership probabilities for  $2 \times 10^6$  objects
- Constructed from  $10^5$  objects (missing values included)
- Non-homogeneous uncertainties propagated from all observables
- Full posterior distribution of the cluster
- Partial results submitted to A&A
  1. The Seven Sisters DANCe III: Bayesian Hierarchical Model
  2. The Seven Sisters DANCe IV: Projected Spatial Distribution

# Limitations

- Field model
  - Does not include uncertainties
  - It is held fixed → simultaneously inferred with the cluster
- Missing values
  - Not at random! → We need the detection probability
- Computing time → Parallelism in GPUs

# Perspectives

- Apply to other clusters → COSMIC-DANCe → Luminosity distributions of Nearby Clusters
- Parallax, Radial velocities, Extinction, Binaries
- Data from: Pan-STARRS, Gaia, LSST, etc.
- Use empirical isochrones to improve stellar models.
- Liberate the software